Estimating down deadwood from FIA forest inventory variables in Maine

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"Capsule": *The down deadwood component of forest carbon can be modeled at the regional scale as a function of stand size class, basal area of dead and cut trees.*

Abstract

Down deadwood (DDW) is a carbon component important in the function and structure of forest ecosystems, but estimating DDW is problematic because these data are not widely available in forest inventory databases. However, DDW data were collected on USDA Forest Service Forest Inventory and Analysis (FIA) plots during Maine's 1995 inventory. This study examines ways to predict DDW biomass from other FIA variables so that DDW could be estimated without tedious measurement. Our results include a regression model that predicts DDW as a function of stand size class, basal area of dead and cut trees, and dummy variables for forest type and forest industry ownership. We also found DDW similar to FIA's standing-tree mortality at a statewide scale. Published by Elsevier Science Ltd.

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1. Introduction

Increasing concentration of atmospheric greenhouse gases has led many nations to seek ways to mitigate greenhouse gas emissions. Forest management is one of the few activities that can remove large amounts of a major greenhouse gas, carbon dioxide, from the atmosphere at relatively low cost. In the United States, about 210 Tg (10^{12} g) of carbon—11% of all United States greenhouse gas emissions reported in 1998—is sequestered annually in forests and forest products (Birdsey and Heath, 1995; USEPA, 2000). Therefore, accurate measurement of carbon in forests is essential.

The most reliable and verifiable method of estimating carbon at a national scale in United States forests is conversion of forest inventory data to carbon (Birdsey, 1992) or linking inventory data with additional information from models (Heath and Birdsey, 1993; Plantinga and Birdsey, 1993; Turner et al., 1995). These methods generally partition the forest ecosystem into

carbon pools for live trees, down deadwood (DDW), standing dead trees, understory vegetation, forest floor material, and soil carbon.

Estimating the DDW or downed coarse woody debris pool has been particularly problematic. Down deadwood is not consistently available from the largest inventory data source, the US Department of Agriculture, Forest Service, Forest Inventory and Analysis (FIA) program, and there is little data on DDW in the scientific literature. Plantinga and Birdsey (1993) dealt with DDW in a carbon assessment by assuming logging residue immediately decayed and its carbon was emitted into the atmosphere, while Turner et al. (1995) used a modeling approach to estimate DDW that was based on the work of Harmon (1993).

In this study, we investigated methods to predict biomass of DDW from FIA data in Maine, which includes a special collection of DDW. This work will be useful for improving methodology to track DDW in forest carbon budgets and to provide information for other uses of DDW. Besides carbon sequestration, DDW is important in understanding energy flows, nutrient cycling, and soil movement, and for serving as habitat for numerous organisms (McMinn and Crossley, 1996).

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Table 1
Maine's FIA forest types grouped into 6 categories

Forest type group	FIA forest type and code number
Northern white cedar/eastern hemlock	White pine/hemlock (4), hemlock (5), northern white cedar (14), tamarack (15)
Spruce/fir	Balsam fir (11), black spruce (12), red spruce/fir (13), white spruce (16), Norway spruce (17), red spruce (19)
Pine/oak	Jack pine (1), red pine (2), white pine (3), pitch pine (38), oak/pine (40), white pine/red oak (41), other oak/pine (49), post oak/black oak/bear oak (51), chestnut oak (52), white oak/red oak/hickory (53), white oak (54), northern red oak (55), mixed central hardwoods (59)
Aspen/birch	Aspen (91), paper birch (92), gray birch (93)
Sugar maple/beech/birch	Sugar maple/beech/yellow birch (81), mixed northern hardwoods (89)
Red maple and other hardwood	Sugarberry/American elm/green ash (63), black ash/American elm 'red maple (71), willow (74), red maple/lowland (76), black cherry (82), red maple/northern hardwood (84), red maple/upland (87), northern hardwood/old field (88), nonstocked (99)
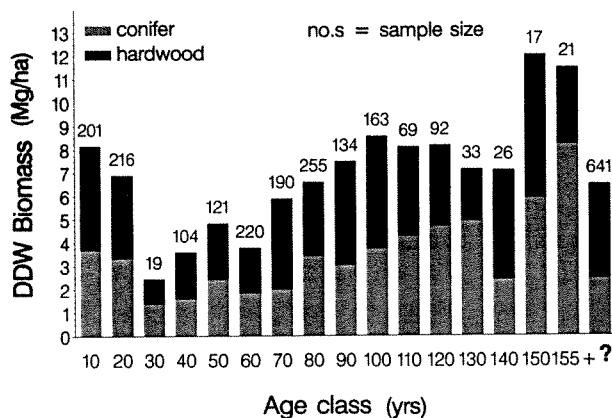
Fig. 2. Slight u-shaped pattern for down deadwood biomass in relation to stand age for Maine FIA plots. Age is missing (?) for 641 plots.

included an average age calculated from, at most, three trees per plot, and these trees were primarily selected for site index determination.

More general descriptors of stand structure were sought to complement or replace stand age. FIA data included a stand size class variable for seedling, pole, and saw timber stand sizes, but the basic tree measurements also provided opportunity to calculate more stand structure variables. Tree density, average size, and species composition were estimated by the suite of variables—basal area, trees per area, stand density index (Long, 1998), and quadratic mean diameter—calculated separately for hardwoods and conifers divided into live, dead, and cut classes. Calculations included all trees in each class, and were repeated twice to include trees larger than threshold diameters of 12.7 and 25.4 cm dbh. The smaller threshold corresponded to the plot size for trees less than 12.7 cm dbh, which were sampled on five subplots with a total area 12 times smaller than plot

size for larger trees. Overall, the measurements and subsequent calculations provided about 300 variables for analysis.

All variables were evaluated for predicting DDW by using three methods: (1) stepwise regression, (2) classification and regression trees (CART), and (3) multivariate adaptive regression splines (MARS). The stepwise regression analysis was limited to continuous variables and to subsets of more or less independent variables. On the other hand, all variables were examined simultaneously in the nonparametric CART and MARS data mining tools (Salford Systems, 2000). Because the dependent variable (DDW) was continuous, MARS was theoretically more appropriate than CART, but both gave similar results. CART and MARS are classification regression-tree methods where the most important variables identify data splits at the top of dichotomous tree diagrams. These nonparametric methods require few assumptions, but results are somewhat dependent on which options are used in the software. CART and MARS were used in an exploratory mode to find the most important "splitter variables" for a range of options. The regression-tree and stepwise-regression results were compared with select variables consistently important among all methods.

3. Results and discussion

From the variable selection analyses, we found DDW biomass most related to variable functions of dead standing trees and cut tree stumps. For example, basal area, biomass, stand density index, and numbers of standing dead and cut trees were consistently important in CART and MARS data mining results. Live tree variables, on the other hand, showed little relationship to DDW. We chose basal areas of standing dead and

cut trees as simple and adequate summary variables of the CART and MARS findings.

None of the procedures showed much value in stand age for estimating DDW. This was contrary to previous findings, but it could have been due to inadequate FIA age measurement procedures. Therefore, a surrogate for age was sought because we wanted to include effects of stand development over time in our model. Stand size class categorized by forest type showed the typical u-shaped pattern others have observed for age data (Fig. 3).

Model refinements on fitting DDW to a function of dead/cut basal area and stand size class showed some value for including some forest types and ownership as dummy variables. Regression was used to parameterize our model:

$$DDW = 8.9096 + 1.3169 \times d_i + 1.2795 \times d_s$$
$$+ 1.6342 \times d_m - 6.9071 \times X_1 + 1.6361$$
$$\times X_1^2 + 0.7042 \times X_2 + 0.2260 \times X_3 \quad (1)$$

where: DDW = down deadwood biomass > 7.6 cm diameter (Mg/ha); $d_i = 1$ if forest industry ownership, 0 otherwise; $d_s = 1$ if forest type spruce/fir, 0 otherwise; $d_m = 1$ if forest type sugar maple/beech/birch, 0 otherwise; X_1 = FIA stand size class (3 = seedling, 2 = pole-timber, 1 = sawtimber); X_2 = basal area of standing dead trees 7.6-cm dbh and larger (m²/ha); X_2 = basal area of standing dead trees 7.6-cm dbh and larger (m²/ha); X_3 = basal area of recently cut trees 7.6-cm dbh and larger (m²/ha).

Although the data fit the model rather poorly $(R^2 = 0.10)$, the large sample size ($n = 2522$) randomly distributed across Maine ensured unbiased predictions

for the State. Another way to view the model is as a slight improvement over estimating DDW from a table of mean values within forest type, stand size, and owner.

Because the model is empirical, the Maine forest types are unlikely to extrapolate to other states. Therefore, a less detailed hardwood and conifer forest type version was also fit to the data:

$$DDW = 10.0666 + 1.6102 \times d_i - 0.4344 \times d_{cr}$$
$$- 7.1780 \times X_1 + 1.6924 \times X_1^2 + 0.7108$$
$$\times X_2 + 0.2190 \times X_3 \quad (2)$$
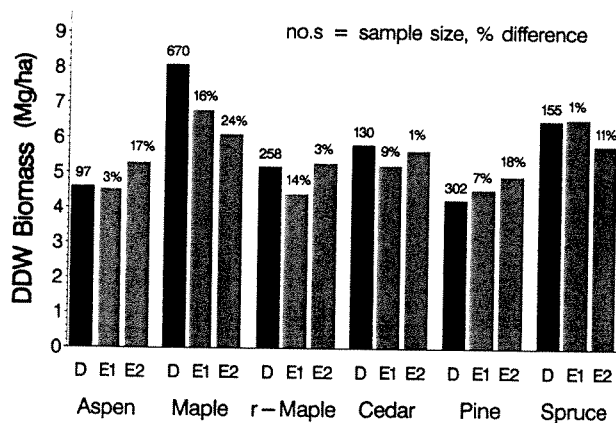
Fig. 4. Down deadwood data (D) from New Hampshire and Vermont compares favorably with predictions from our Eqs. (1) and (2) (E1, E2) for most forest type groups (aspen/birch, maple/beech/birch, red maple/other hardwoods, cedar/hemlock, pine/oak, spruce/fir).
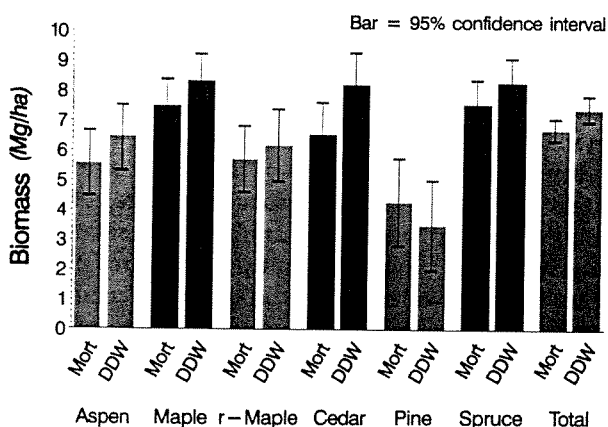
Fig. 5. Biomass of standing mortality trees (Mort) that have died since the last FIA inventory (14 years ago) is comparable with down deadwood (DDW) in Maine for forest type groups (aspen/birch, maple/beech/birch, red maple/other hardwoods, cedar/hemlock, pine/oak, spruce/fir). The 95% confidence interval is based on variance calculation from the double sampling for stratification design. Sample size was reduced to 1848 because FIA does not account for mortality on newly established plots.
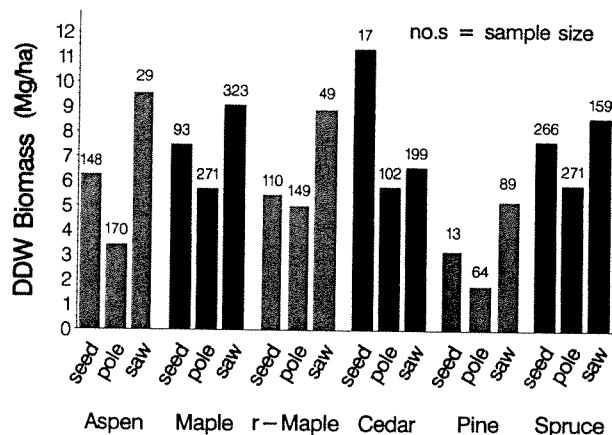

Fig. 3. Down deadwood is related to stand size class (seedling, pole-timber, sawtimber) in u-shaped pattern for forest type groups (aspen/birch, maple/beech/birch, red maple/other hardwoods, cedar/hemlock, pine/oak, spruce/fir).

where $d_{cr} = 1$ if conifer forest type, 0 otherwise.

Validation against similar data for Vermont and New Hampshire showed reasonable results for the Eq. (1) model. Down deadwood for most forest types was underestimated from 3 to 16% but pine and spruce/fir were overestimated by 7 and 1%, respectively (Fig. 4). The more general Eq. (2), lacking forest type detail, produced similar but slightly more variable results. Because the model is empirical, greater difference should be expected for application to States farther from Maine.

In comparing DDW biomass with other components of forest biomass, we noticed biomass of FIA's mortality trees was closely related to DDW biomass at the Statewide scale for Maine (Fig. 5). FIA mortality included all standing trees (over 12.6-cm dbh) that had died during the 14-year period since the last inventory. This relationship was not evident on a plot-by-plot basis, but only at the State level. Vermont and New Hampshire data also showed the same curious pattern. Although surprising, this warrants further study because standing crop DDW biomass may be in balance with disturbance factors in some manner where DDW could be modeled as a function of dead tree inputs and decay rates in a very general sense at a broad geographic scale.

n summary, DDW can be estimated from FIA variables—dead and cut tree basal area, stand size class, forest type, and ownership. Because our model is empirically derived and lacking theoretical justification for application outside Maine, its best use might be in a regression subsampling strategy where new parameters are estimated for each State inventoried by FIA. Also, biomass of FIA mortality trees is a point of comparison for our model. For Maine, Vermont, and New Hampshire, the DDW biomass and standing mortality biomass were quite similar.

Acknowledgements

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